

Barotropic Energy Conversion During Indian Summer Monsoon: Implication of Central Indian Ocean Mode Simulation in CMIP6

Jianhuang Qin

Hohai University

Lei Zhou (✉ zhoulei1588@sjtu.edu.cn)

Shanghai Jiao Tong University <https://orcid.org/0000-0002-0433-3991>

Ze Meng

Shanghai Jiao Tong University

Baosheng Li

State Key Laboratory of Satellite Ocean Environment Dynamics

Tao Lian

State Key Laboratory of Satellite Ocean Environment Dynamics

Raghu Murtugudde

University of Maryland at College Park

Research Article

Keywords: central Indian Ocean (CIO), Dynamic diagnostics, CMIP6, MISO

Posted Date: July 27th, 2021

DOI: <https://doi.org/10.21203/rs.3.rs-715759/v1>

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Version of Record: A version of this preprint was published at Climate Dynamics on January 11th, 2022.
See the published version at <https://doi.org/10.1007/s00382-021-06087-y>.

1 **Barotropic energy conversion during Indian summer monsoon:**
2 **Implication of Central Indian Ocean Mode Simulation in CMIP6**

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4 Jianhuang Qin^{1,2}, Lei Zhou^{3,4*}, Ze Meng³, Baosheng Li², Tao Lian^{2,3,4},

5

Raghu Murtugudde⁵

6 *1. College of Oceanography, Hohai University, Nanjing, China*

7 *2. State Key Laboratory of Satellite Ocean Environment Dynamics, Second Institute of*
8 *Oceanography, Ministry of Natural Resources, Hangzhou, China*

9 *3. School of Oceanography, Shanghai Jiao Tong University, Shanghai, China*

10 *4. Southern Marine Science and Engineering Guangdong Laboratory (Zhuhai)*

11 *5. University of Maryland, College Park, Maryland, USA*

12

13 *Corresponding author: Lei Zhou (zhoulei1588@sjtu.edu.cn)*

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16

17 **Abstract**

18 The simulation and prediction of the Indian summer monsoon (ISM) and its
19 intraseasonal component in climate models remain a grand scientific challenge for
20 models. Recently, an intraseasonal mode was proposed over the tropical Indian Ocean,
21 named central Indian Ocean (CIO) mode. The CIO mode index and with monsoon
22 intraseasonal oscillations (MISO) have a high correlation. In this study, the simulations
23 of the CIO mode in the sixth phase of the Coupled Model Intercomparison Project
24 (CMIP6) models are examined. Although the coupled ocean-atmosphere feedbacks
25 associated with the CIO mode are not fully reproduced, the results show that a better
26 depiction of the CIO mode in CMIP6 models is favorable for a better simulation of
27 northward-propagating MISO and heavy rainfall during the ISM. Dynamic diagnostics
28 unveil that the rendition of the CIO mode is dominated by kinetic energy conversion
29 from the background to the intraseasonal variability. Furthermore, kinetic energy
30 conversion is controlled by the meridional shear of background zonal winds ($\frac{\partial \bar{u}}{\partial y}$), which
31 is underestimated in most CMIP6 models, leading to a weak barotropic instability. As
32 a result, a better simulation of $\frac{\partial \bar{u}}{\partial y}$ is required for improving the CIO mode simulation
33 in climate models, which helps to improve the simulation and prediction skill of
34 northward-propagating MISO and monsoonal precipitation.

38 **1. Introduction**

39 The Indian summer monsoon (ISM) precipitation has tremendous scientific and
40 socioeconomic significance, which contributes about 80% of the total annual
41 precipitation over the Indian subcontinent (Bollasina, 2014) and has a substantial
42 influence on agricultural and industrial productions. The ISM precipitation has two
43 significant timescales; one is between 30 and 60 days [known as intraseasonal
44 variability; Yasunari, 1981], and the other one is between 10 and 20 days (i.e., quasi-
45 biweekly variability; Chatterjee and Goswami, 2004). The former is controlled by the
46 monsoon intraseasonal oscillation (MISO; Goswami 2005; Shukla 2014), which
47 dominates the active and break spells in monsoonal precipitation. MISO can explain
48 approximately 60% of total precipitation variance over the Bay of Bengal (BoB)
49 (Goswami 2005; Waliser 2006; Shukla 2014). To date, the simulation and prediction of
50 monsoonal precipitation and MISO remains a great challenge for contemporary models
51 (e.g., Sabeerali et al. 2013; Wang et al. 2015; Goswami and Chakravorty, 2017; Hazra
52 et al. 2017). The predictability of ISM is dependent on its close relationship with the El
53 Niño–Southern Oscillation (ENSO, e.g., Gill et al. 2015), the Atlantic Niño
54 (Pottapinjara et al. 2014) and the Indian Ocean Dipole/Zonal Mode (IODZM, e.g.,
55 Murtugudde et al. 2000; Ashok et al. 2001). However, the intraseasonal variabilities are
56 considered as a “desert of predictability” for a long time (Waliser et al. 2003; Vitart et
57 al. 2017). Thus, insight into intraseasonal variabilities over the tropical Indian Ocean
58 can help to facilitate a better simulation of the ISM, and to advance predictive
59 understanding of the ISM precipitation.

60 Recently, an intraseasonal mode, i.e., the central Indian Ocean (CIO) mode, was
61 proposed by Zhou et al. (2017a). It is proved that the CIO mode is closely related to
62 MISO propagation and monsoonal precipitation. The CIO mode is obtained by the first
63 combined Empirical Orthogonal Function (EOF) mode of intraseasonal sea surface
64 temperatures (SSTs) and intraseasonal 850 hPa zonal winds (referred to as U850
65 hereafter) over the tropical Indian Ocean, and the corresponding principal component
66 (PC) is referred to as the CIO mode index. The CIO mode is not sensitive to the
67 reanalysis products and spatiotemporal domains (Zhou et al. 2017a, 2018; Qin et al.
68 2020). The spatial pattern of the positive CIO mode is shown in Fig. 1a during the
69 period of 1998-2014. The intraseasonal SST node of the positive CIO mode captures
70 positive anomalies along the central Indian Ocean, accompanied with an anti-cyclonic
71 gyre in the lower troposphere. The CIO mode is energetic during boreal summer (June-
72 September), which is attributable to the enhanced transmission of kinetic energy from
73 the background state to the intraseasonal timescales (Zhou et al. 2017b). It is verified
74 that the positive CIO mode plays an important role in driving the heavy precipitation
75 during the ISM, via changing the propagation direction of the intraseasonal oscillations
76 (Zhou et al. 2017a; Qin et al. 2020). The positive CIO mode facilitates the transition
77 from the eastward-propagating intraseasonal variabilities [commonly known as
78 Madden–Julian oscillation (MJO); Madden and Julian 1971, 1972; Zhang 2005] to the
79 northward-propagating component during the ISM, since the easterly vertical wind
80 shear associated with the positive CIO mode are favorable for the latter (Fu et al. 2004;
81 Jiang et al. 2004; Kang et al. 2010; Zhou and Murtugudde 2014). As a result, the

82 positive CIO mode shows a significant positive correlation with intraseasonal
83 precipitation over the BoB (Fig. 1b), where the total rainfall and intraseasonal rainfall
84 variance are large (Fig. 1c and d). It is also argued that the relationships of the CIO
85 mode with MISO and ISM are independent on ENSO and IODZM (Zhou et al. 2017b),
86 which indicates that the CIO mode can provide an independent way to improve the
87 simulation of MISO and monsoonal precipitation.

88 The evaluation of CIO mode simulations was investigated in the Community Earth
89 System Model (CESM) and Subseasonal-to-Seasonal (S2S) air-sea coupled models
90 (Zhou et al. 2018; Qin et al. 2020). A consistent conclusion was that a better depiction
91 of the CIO mode in a model tends to a better reappearance of northward-propagating
92 MISO and heavier intraseasonal rainfall during the ISM. However, the CIO mode is not
93 well captured in CESM. As a result, the simulated monsoonal precipitation and the
94 northward-propagating MISO are weaker than observed. Although most of the S2S air-
95 sea coupled models can reproduce the CIO mode on initial days, the simulations of the
96 CIO mode become deficient rapidly as the lead time for forecast increases. Such biases
97 in the CIO mode simulation are mainly attributed to the weak meridional shear of the
98 low-frequency zonal winds (a low-pass filter of 100 days) in the above climate models,
99 which reduces the barotropic kinetic energy conversion from the background state to
100 intraseasonal variabilities. Hence, the CIO mode is not strong enough to reinforce the
101 moisture loading in the subtropical mid-level troposphere, which benefits the northward
102 propagation of MISO. In addition, the intraseasonal meridional wind related to the CIO
103 mode is also important for the moisture transport over the BoB in S2S air-sea coupled

104 models.

105 Nevertheless, the evaluations of the CIO mode simulation were only based on
106 CESM and 6 air-sea coupled models in the S2S database. More state-of-the-art climate
107 models are needed to examine the simulation of the CIO mode as well as its impacts on
108 the simulation of MISO and the ISM. The Coupled Model Intercomparison Project
109 (CMIP), which began in 1995 under the auspices of the World Climate Research
110 Programme (WCRP), is now in its sixth phase (CMIP6). Some studies have
111 suggested that the CMIP6 models yield better simulations of MJOs than the CMIP5
112 models, such as a slower eastward propagation, stronger teleconnection pattern, and
113 longer persistence of MJOs (e.g., Ahn et al. 2020; Wang et al. 2020). The inter-
114 comparisons among different CMIP6 models help to promote the process
115 understandings of the CIO mode and are expected to improve the simulations of MISO
116 and monsoonal precipitation. This is the motivation for this paper. The remainder of
117 this paper is organized as follows. Section 2 introduces model configurations and the
118 methods used in this study. In Section 3, the assessments of the CIO mode and related
119 processes simulations are examined. Finally, the summary and discussion are shown in
120 Section 4.

121

122 **2. Data and methods**

123 The simulated daily atmospheric and oceanic data (including SST, winds,
124 precipitation and specific humidity) are retrieved from the Earth System Grid (ESG)
125 data portal for 18 CMIP6 models (<https://esgf-node.llnl.gov/search/cmip6/>), which are

126 from the Historical experiments during the period of 1998-2014 (the time range for the
127 study). The Historical experiments represent present-day climate, and only the first
128 ensemble member (r1i1p1f1) is analyzed from each CMIP6 model. The general
129 information of the 18 CMIP6 models, including model resolutions and model numbers,
130 are listed in Table 1. Before the inter-comparison among different models, all variables
131 are interpolated to a horizontal resolution of 0.5° latitude \times 0.5° longitude, which has
132 no impacts on the extraction of intraseasonal variabilities in this study (not shown).

133 Daily SST data with a resolution of 0.25° latitude \times 0.25° longitude from 1998 to
134 2014 are obtained from the National Oceanic and Atmospheric Administration (NOAA)
135 Optimum Interpolated SST (OISST; Reynolds et al. 2007). Atmospheric variables are
136 obtained from the European Center for Medium-Range Weather Forecast (ECMWF)
137 ERA5 reanalysis dataset (Hersbach et al. 2019), including winds and specific humidity,
138 with horizontal resolution of $0.25^\circ \times 0.25^\circ$ and temporal resolution of 6 hours during
139 the period of 1998-2014. Daily precipitation during the same period is the 3B42 product
140 from the Tropical Rainfall Measuring Mission (TRMM) rainfall data (Kummerow et al.
141 2000). All intraseasonal oscillations in both simulations and observations are obtained
142 with a 20–100-day band-pass Butterworth filter. Student's t -test on the basis of a
143 difference between sample means is used to test the statistical significance of the
144 correlation coefficient.

145 The projection method is adopted to acquire the CIO mode index in CMIP6 models.
146 The projected CIO mode index is calculated as

147
$$A(x, y, t) = B(x, y) \cdot index(t) + R1 \dots (1)$$

148 where A is the three-dimensional data including intraseasonal SST and U850
149 anomalies; B represents the spatial structure of the observed CIO mode; $R1$ is the
150 residual term; x , y and t are the number of latitude, longitude and time, respectively.

151

152 **3. Results**

153 *3.1 Simulation of the Mean Climate State and the Monsoon*

154 Figure 2 shows the difference of mean SST between observations and simulations
155 in CMIP6 models (simulations minus observations) during boreal summer (June-
156 September). The differences are less than 2°C in the tropical Indian Ocean. The Indian
157 Ocean warm pool exists along the equator from the central to the eastern Indian Ocean
158 (contours in every panel in Fig. 2), but they are underestimated in CMIP6 models.
159 Conversely, the simulated SSTs in the western Indian Ocean are warmer than the
160 observations. The exceptions are MPI-ESM-1-2-HAM, MPI-ESM1-2-LR and NESM3
161 (#11, 13 and 15, Fig 2k, m and o), in which the mean simulated SSTs are colder than
162 the observations over the entire tropical Indian Ocean during boreal summer. The
163 differences of the simulated mean U850 from the observations are shown in Fig. 3. The
164 U850 in reanalysis consist westerlies (easterlies) in the north (south) of equator over
165 the tropical Indian Ocean during boreal summer (contours). Although the bias of U850
166 is less than 5 m s^{-1} , almost all models represent stronger westerlies over the BoB (except
167 IPSL-CM6A-LR, #9, Fig. 3i) and weaker westerlies over the Arabian sea. The weak
168 westerly biases reduce upwelling in the western basin, probably leading to the warm
169 SST bias in models (Fig. 2). Moreover, the easterlies along the equator are weaker than

170 those in observations, leading to reduced warm water convergence to the warm pool.

171 As a result, the SSTs on the equator are colder in CMIP6 models.

172 The mean precipitation in observations reach the maximum over the BoB (Fig. 1c)
173 during boreal summer (from June to September). The standard deviation (STD) of
174 intraseasonal rainfall is also large at the same locations (Fig. 1d). As shown in Fig.4,
175 the mean precipitation and the STD of intraseasonal precipitation averaged within 10°-
176 20°N and 80°-100°E during boreal summer are 13.77 mm day⁻¹ (red line) and 9.4 mm
177 day⁻¹ (gray line) in TRMM, respectively. Similarly, the red and gray bars in Fig. 4
178 represent the mean precipitation and the STD of intraseasonal precipitation calculated
179 using CMIP6 models, respectively. One can see that the precipitation and its
180 intraseasonal variability are much weaker in all CMIP6 models than in nature. It
181 indicates that the underestimated monsoonal precipitation remains a persistent problem
182 for most climate models. Due to the close relationship between the CIO mode and ISM,
183 it can be reasonably assumed that the simulated CIO mode is not strong enough in
184 CMIP6 models. Therefore, there are likely to be some inadequacies in the simulation
185 of CIO mode and its processes, which are discussed in more detail below.

186

187 *3.2 Evaluation of Simulated CIO Mode*

188 The CIO mode is defined as the first combined EOF mode between intraseasonal
189 SST anomalies and intraseasonal U850 anomalies. Figure 5 shows the simulated CIO
190 mode calculated by CMIP6 model outputs. The variance explanations of the simulated
191 CIO mode range from 7.0% to 10.4% in CMIP6 models (around 9.2% in ERA5

192 reanalysis, Fig. 1a). In nature, the anti-cyclone and corresponding downdraft enhance
193 the easterly vertical wind shear, which benefits the northward propagation of MISO
194 (Zhou et al. 2017a). Meanwhile, the downdraft over the central Indian Ocean increases
195 the incident solar radiation at the sea surface, leading to warm SST anomalies during
196 the positive CIO mode. It suggests that the atmosphere plays an active role in the ocean-
197 atmosphere interaction (Xi et al. 2015). In the simulations, the easterlies along the
198 equator associated with the positive phase of CIO mode are well captured in almost all
199 CMIP6 models (contours in Fig. 5), except ACCESS-ESM1-5 and GFDL-CM4 (#2, 8,
200 Fig. 5b and 5h), which are dominated by westerlies along the equator. However,
201 westerly wind anomalies, which in reality are around 10°N, are barely evident in either
202 simulation. Compared with observations and the ERA5 reanalysis (Fig. 1a), the
203 simulated westerly wind anomalies shift to the southern hemisphere in CMIP6 models.
204 As a result, the warm SST anomalies in the central Indian Ocean are quite realistic in
205 CMIP6 models (colors in Fig. 5). The mismatch between the simulations and
206 observations reveals the inadequacy of the ocean-atmosphere coupling, which is a
207 typical shortcoming of monsoon simulations as emphasized by many previous studies
208 (e.g., Meehl et al. 2012; Goswami et al. 2014).

209 Given the misrepresentation of SSTs in the simulated CIO mode using the EOF
210 analysis, another way to evaluate the simulated CIO mode is to project the model
211 outputs onto the observed CIO mode (Fig. 1a). The purpose of the projection method is
212 to estimate how much actual CIO mode is captured in each CMIP6 model simulation.
213 The projected CIO mode index in CMIP6 models is calculated by Eq. (1). Figure 6

214 shows the correlation coefficients of intraseasonal precipitation with the projected CIO
215 mode index during boreal summer (from June to September) in each CMIP6 model. It
216 is obvious that the correlations over the BoB are significantly positive with a maximum
217 over 0.5 in all CMIP6 models, which emphasizes the close relationship between the
218 CIO mode and monsoonal precipitation.

219 The STDs of projected CIO mode index during the ISM (from June to September)
220 are shown in the x-axis of Fig. 7, which represents the intensity of the simulated CIO
221 mode in each CMIP6 model. The y-axis in Fig. 7 represents the strength of
222 intraseasonal precipitation over the BoB (10° - 20° N, 80° - 100° E, the same as gray bars
223 in Fig. 5). The correlation coefficient between the STD of intraseasonal precipitation
224 and the STD of projected CIO mode index is 0.71 (significant at a 99% confidence
225 level). This result also indicates that a pronounced CIO mode is helpful to enhance the
226 monsoonal precipitation, which is lacking in most CMIP6 models. Particularly,
227 CESM2-FV2, CESM2-WACCM-FV2, MIROC6 and SAM0-UNICON (#5, 6, 10 and
228 18, considered as the well simulated group, listed in Tab. 2) show better simulations of
229 monsoonal precipitation and the CIO mode than the other models do, and all of them
230 reproduce higher mean precipitation and stronger intraseasonal precipitation (red and
231 gray bars in Fig. 5). Moreover, the correlations of intraseasonal precipitation with the
232 projected CIO mode index in the well simulated group are also higher than other models
233 (Fig. 6d, e, j and r), and more similar to that in observations (Fig. 1b). In contrast, the
234 STDs of the CIO mode are weak in ACCESS-CM2, ACCESS-ESM1-5, CanESM5 and
235 IPSL-CM6A-LR (#1, 2, 3 and 9, considered as the poorly simulated group, listed in

236 Tab. 2). This leads to weak STD of monsoonal precipitation and lower correlations with
237 intraseasonal precipitation over the BoB (Fig. 6a, b, c, and i) during the ISM.

238 Furthermore, the energetics of MISO are examined from the equator to 30°N
239 during the positive CIO mode events (Zhou et al. 2017a; Qin et al. 2020). According to
240 the CIO mode index, the positive CIO mode events are defined by a local maximum
241 and larger than its STD during boreal summer (June-September). There are 54 positive
242 CIO mode events using ERA5 reanalysis and OISST from 1998 to 2014. The numbers
243 of positive CIO mode events using the projected CIO mode index in each CMIP6 model
244 are listed in Tab. 3. Almost all CMIP6 models represent slightly more positive CIO
245 mode events (ranges from 52 to 68) than observations. The northward MISO
246 propagation can be clearly seen in the Hovmöller diagrams of intraseasonal zonal winds
247 (colors), intraseasonal outgoing longwave radiation (OLR; white contours), and
248 intraseasonal rainfall (black contours) averaged between 80°E and 90°E in most CMIP6
249 models (along the green reference lines, Fig. 8). Day 0 on the x-axis is the day when
250 the projected CIO mode index peaks during the ISM. In comparison, the speed of
251 northward propagation has no obvious difference (approximate 1° day^{-1}) among
252 different models. The maximum of intraseasonal precipitation is higher than 5 mm day^{-1}
253 ¹ with negative OLR anomalies below 15 W m^{-2} and can reach up to 30°N in the well
254 simulated group (#5, 6, 10, and 18; Fig. 8e, f, j and r) when the projected CIO mode
255 index peaks. However, the intraseasonal zonal winds, precipitation and OLR are much
256 weaker in the poorly simulated group (#1, 2, 3 and 9; Fig. 8a, b, c and i), and are
257 restricted to the south of 20°N. Therefore, it can be concluded that a better simulation

258 of the CIO mode benefits the simulations of northward-propagating MISO and
259 monsoonal precipitation during the ISM.

260

261 *3.3 Dynamics of the CIO mode in CMIP6 models*

262 Figure 9 shows the composite of simulated intraseasonal specific humidity and
263 vertical velocity averaged between 80°E and 90°E on the peak days of the CIO mode
264 index. The intraseasonal specific humidity (colors) shows negative anomalies over the
265 equator and positive anomalies to the north of 20°N and upward to the mid-troposphere
266 in all CMIP6 models. These conditions together with the vertical velocity (contours)
267 play an important role in generating precipitation. Positive specific humidity anomalies
268 and upward motions induce the heavy rainfall and latent heat release between 10°N and
269 25°N. However, the upward motions are not well captured in all CMIP6 models (dashed
270 contours in Fig. 9). The center of upward motions hardly reaches the north of 20°N in
271 most CMIP6 models. Particularly, they are weak in the poorly simulated group (#1, 2,
272 3 and 9; Fig. 9a, b, c and i), leading to reduced rainfall over the BoB. In contrast, the
273 positive intraseasonal specific humidity anomalies are aligned with the negative
274 intraseasonal omega in the well simulated group (#5, 6, 10 and 18; Fig. 9e, f, j and r),
275 which contribute to heavy rainfall from 10°N to 30°N where the monsoonal
276 precipitation is large. This result suggests that the bias of winds is larger than the bias
277 of specific humidity in CMIP6 models, although winds and heat sources can be related
278 to each other in a coupled system.

279 Previous studies have reported the importance of kinetic energy (KE) conversion

280 during the CIO mode (Zhou et al. 2017b; 2018; Qin et al. 2020). The kinetic energy
 281 budget is checked for the CMIP6 models. All variables are decomposed into three
 282 components. For instance, the zonal wind is decomposed as $u = \bar{u} + u' + u''$, where
 283 \bar{u} is obtained with a low-pass filter of 100 days representing the background state of
 284 zonal wind, u' is the intraseasonal zonal wind, and u'' (obtained with a high-pass
 285 filtering of 20 days) is the high-frequency variability. Following Zhou et al. (2012), the
 286 budget of intraseasonal kinetic energy (KE') is written as

$$287 \quad \frac{\partial KE'}{\partial t} = -\bar{\mathbf{V}} \cdot \nabla KE' + [KE' \times \overline{KE}] + [KE' \times KE''] + [KE' \times PE'] - \nabla(\mathbf{V}' \cdot \Phi')$$

$$288 \quad + R2 \dots (2),$$

289 where \mathbf{V} , Φ and PE represent the horizontal wind (including zonal and meridional
 290 winds), geopotential and potential energy, respectively; $-\bar{\mathbf{V}} \cdot \nabla KE'$ is the advection
 291 term of intraseasonal kinetic energy; $[KE' \times \overline{KE}]$ represents the kinetic energy
 292 conversion between the background and intraseasonal variabilities; $[KE' \times KE'']$ is
 293 the conversion between the intraseasonal kinetic energy variabilities and higher
 294 frequency oscillations; $[KE' \times PE']$ represents the energy conversion between
 295 intraseasonal kinetic energy and potential energy; $-\nabla(\bar{\mathbf{V}}' \cdot \Phi')$ is the work done by the
 296 pressure gradient force; and R2 is the residual term. More details of the kinetic energy
 297 budget can be seen in Zhou et al. (2012).

298 In ERA5 reanalysis, pronounced intraseasonal kinetic energy occurs in three
 299 regions, i.e., the Indian Peninsula to BoB, Arabian Sea, and the central Indian Ocean
 300 around the equator (Fig. 10a). The former two are related to the ISM, and the last one
 301 is associated with the CIO mode. These results are consistent with that calculated by

302 ERA-Interim data and NCEP Reanalysis 2 (Zhou et al. 2017b; Zhou et al. 2018).
 303 According to the kinetic energy budget analysis (Eq. 2), the kinetic energy on
 304 intraseasonal timescales is provided by $[KE' \times \overline{KE}]$ during the ISM, while other terms
 305 in Eq. (2) are generally small. As shown in Fig. 10b, positive $[KE' \times \overline{KE}]$ represents
 306 kinetic energy conversion from the background to the intraseasonal variabilities, which
 307 boosts the kinetic energy on intraseasonal timescales. In the simulations, the
 308 intraseasonal kinetic energy and $[KE' \times \overline{KE}]$ at 850 hPa averaged during the ISM in
 309 CMIP6 models are shown in Fig. 11 and Fig. 12, respectively. A noticeable bias is that
 310 the pronounced intraseasonal kinetic energy over the central Indian Ocean is largely
 311 missing in all CMIP6 models, due to the weaker $[KE' \times \overline{KE}]$ along the equator.
 312 However, the intraseasonal kinetic energy center associated with strong $[KE' \times \overline{KE}]$
 313 over the BoB is well captured in most CMIP6 models, especially in the well simulated
 314 group (#5, 6, 10 and 18; Fig. 11e, f, j and r). In contrast, the intraseasonal kinetic energy
 315 over the BoB is smaller than 5 J kg^{-1} in the poorly simulated group (#1, 2, 3 and 9; Fig.
 316 11a, b, c and i), in which the simulations of the ISM are weak. Moreover, $[KE' \times \overline{KE}]$
 317 is also weak (smaller than $2 \text{ J day}^{-1} \text{ kg}^{-1}$) in the poorly simulated group (#1, 2, 3 and 9;
 318 Fig. 12a, b, c and i). As a result, the intraseasonal kinetic energy is not strong enough
 319 for capturing the CIO mode and the monsoonal precipitation.

320 Previous studies have demonstrated that the kinetic energy transfer from the
 321 background to the intraseasonal variability is dominated by the barotropic instability of
 322 the background state (Holton and Hakim 2013; Vallis 2017). $[KE' \times \overline{KE}]$ is driven by
 323 the meridional shear of background zonal winds ($\frac{\partial \bar{u}}{\partial y}$) during the ISM. In the reanalysis

324 (Fig. 10c), $\frac{\partial \bar{u}}{\partial y}$ shows a meridional train of positive and negative values from the
 325 equator up to 30°N (particularly between 70°E and 100°E). Consistently, the changes
 326 in signs can be seen in the meridional gradient of the quasi-geostrophic potential
 327 vorticity (PV, $\frac{dPV}{dy} = \beta - \frac{\partial^2 \bar{u}}{\partial y^2}$, where β is the meridional gradient of the Coriolis
 328 parameter) from the equator to the north. As shown in Fig. 10d, positive and negative
 329 values of $\beta - \frac{\partial^2 \bar{u}}{\partial y^2}$ occur alternatively in the meridional direction within the ISM
 330 region, which is indicative of the necessary condition for the barotropic instability
 331 (Vallis 2017). Conversely, $\frac{\partial \bar{u}}{\partial y}$ is too weak in CMIP6 models (Fig. 13). $\frac{\partial \bar{u}}{\partial y}$ reaches a
 332 positive maximum around the equator and decreases almost monotonically to its
 333 minimum around 20°N. As a result, $\frac{\partial^2 \bar{u}}{\partial y^2}$ in CMIP6 has the same sign in this region and
 334 it is not strong enough to overcome β (Fig. 14). Therefore, the necessary condition for
 335 the barotropic instability over the BoB cannot be satisfied in CMIP6 models. Compared
 336 with the assessment of intraseasonal kinetic energy in CESM and S2S models (zhou et
 337 al. 2018; Qin et al. 2020), it can be surmised that the underestimated barotropic
 338 instability in contemporary climate models is the essential reason for poor simulation
 339 of the CIO mode and monsoonal precipitation.

340

341 **4. Summary and Discussion**

342 The CIO mode has a strong association with the northward propagation of MISO
 343 and monsoonal precipitation during the ISM, via transferring energy and moisture from
 344 the tropics to the subtropical regions. However, previous studies have investigated that
 345 the simulation of CIO mode in current climate models is poor (Zhou et al. 2018; Qin et

346 al. 2020). In this study, we examined the evaluations of the CIO mode simulation in
347 CMIP6 models. The simulated monsoonal rainfall and its variability on intraseasonal
348 timescale are much weaker than observations. Although the mismatch of the coupled
349 relation between the ocean and the atmosphere associated with the CIO mode remains
350 in CMIP6 models, results confirm that a pronounced CIO mode is helpful to reinforce
351 the northward propagation of MISO and to enhance the monsoonal precipitation over
352 the BoB. Probing deeper, the intraseasonal kinetic energy budget analysis revealed that
353 the poor simulations of the CIO mode and its processes are attributable to the
354 misrepresentation of background winds. Weak meridional shear of background zonal
355 winds ($\frac{\partial \bar{u}}{\partial y}$) in CMIP6 models reduces the kinetic energy conversion from the
356 background state to the intraseasonal variabilities. Then, the intraseasonal kinetic
357 energy is not strong enough to raise a CIO mode event in CMIP6 models. Therefore,
358 barotropic instability is underestimated from the equator up to 30°N (particularly
359 between 70°E and 100°E), and is found to be very weak in most current climate models.

360 Our conclusion that better CIO simulation in CMIP6 models is mainly due to the
361 intensity of the barotropic instability is also supported by the recent model
362 intercomparison studies conducted by Zhou et al. (2018) and Qin et al. (2020). They
363 also found that the couple SST - wind relation between the ocean and the atmosphere
364 in those simulations are opposite to that in observations. Such a mismatch of the ocean
365 and the atmosphere is reduced in CMIP6 models. The bias of SSTs in CMIP6 models
366 is attributed to the poor simulation of winds. Although higher resolution in models show
367 improvements, enhancing the model physics suitable for the higher resolution is also

368 essential. Therefore, more attention is needed for improving both dynamic circulation
369 and thermodynamic processes in climate models, which is expected to in turn improve
370 the simulations of MISO and ISM.

371 During the positive CIO mode, the enhanced easterly wind shear over the tropical
372 central Indian Ocean is favorable for driving intraseasonal oscillations (Zhou et al.
373 2017a; Li et al. 2020). The easterly wind shear during boreal summer is found to be
374 well captured in CMIP6 models (Li et al. 2021). Therefore, the intensity of northward
375 propagation of MISO and monsoonal precipitation is controlled by the strength of the
376 CIO mode. Our results provide a clear way forward to complement the MISO with the
377 boreal summer season focus on the CIO mode. Since the barotropic instability condition
378 during boreal summer is not satisfied in current climate models, numerical experiments
379 may provide us with a better way to understand the importance of barotropic instability
380 for the CIO mode generation and the air-sea interactions related to the CIO
381 mode. Further improvements in convection parameterization schemes associated with
382 barotropic instability in models will be helpful for the betterment of MISO and will lead
383 to the improved simulation of monsoon. This is our future goal and the results will be
384 reported elsewhere.

385

386

387 **Acknowledgement**

388 This work is supported by grants from the National Natural Science Foundation
389 of China (42076001), the open fund of State Key Laboratory of Satellite Ocean
390 Environment Dynamics, Second Institute of Oceanography, MNR (No. QNHX2111),
391 and the Fundamental Research Funds for the Central Universities (B210202142). RM
392 gratefully acknowledges the Visiting Faculty at the Indian Institute of Technology,
393 Bombay. All observation data and reanalysis products for this paper are properly cited
394 and referred to in the reference list. TRMM 3B42 product observation data are
395 downloaded from <https://gpm.nasa.gov/data-access/downloads/trmm>. OISST
396 reanalysis are obtained from <https://www.ncdc.noaa.gov/oisst>. ERA5 reanalysis are
397 available at <https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5>.
398 NCEP-NCAR reanalysis are downloaded from
399 <https://www.esrl.noaa.gov/psd/data/gridded/data.ncep.reanalysis.html>.

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517

518 **Table 1.** The selected 18 CMIP6 models used in our study with names, institutions and
 519 horizontal grid resolution of the atmospheric and ocean variables.

No.	Model	Institution name	Average grid resolution (longitude x latitude)	
			Atmosphere	Ocean
1	ACCESS-CM2	Commonwealth Scientific and Industrial Research Organisation (CSIRO), Australia	1.87°×1.25°	1.0°×1.0°
2	ACCESS-ESM1-5		1.87°×1.25°	1.0°×1.0°
3	CanESM5	Canadian Centre for Climate Modelling and Analysis, Environment and Climate Change Canada, BC, Canada	2.8°×2.8°	1.0°×0.62°
4	CESM2	National Center for Atmospheric Research, Boulder, CO, USA	0.9°×1.25°	0.9°×1.25°
5	CESM2-FV2		1.9°×2.5°	1.9°×2.5°
6	CESM2-WACCM-FV2		1.9°×2.5°	1.9°×2.5°
7	EC-Earth3-Veg	Consortium of various institutions from Spain, Italy, Denmark, Finland, Germany, Ireland, Portugal, Netherlands, Norway, the United Kingdom, Belgium, and Sweden	0.7°×0.7°	1.0°×0.62°
8	GFDL-CM4	Geophysical Fluid Dynamics Laboratory, NOAA, Princeton, NJ, USA	1.0°×1.0°	0.25°×0.16°
9	IPSL-CM6A-LR	Institute Pierre Simon Laplace, Paris, France	2.5°×1.25°	1.0°×0.54°
10	MIROC6	Japan Agency for Marine-Earth Science and Technology, Atmosphere and Ocean Research Institute, National Institute for Environmental Studies, and RIKEN Center for Computational Science, Japan	1.4°×1.4°	1.0°×0.70°
11	MPI-ESM-1-2-HAM	Max Planck Institute fur Meteorologie, Forschungszentrum Julich, University of Oxford, Finnish Meteorological Institute, Leibniz Institute for Tropospheric Research, ETH Zurich	1.87°×1.87°	1.52°×0.82°
12	MPI-ESM1-2-HR		0.94°×0.94°	0.45°×0.45°
13	MPI-ESM1-2-LR		1.87°×1.87°	1.4°×0.82°
14	MRI-ESM2-0	Meteorological Research Institute, Tsukuba, Japan	1.1°×1.1°	1.0°×0.5°
15	NESM3	Nanjing University of Information Science and Technology, Nanjing, China	1.87°×1.87°	1.0°×0.62°
16	NorESM2-LM	Norwegian Climate Centre, Norway	0.467°×1.0°	1.875°×2.5°
17	NorESM2-MM		0.467°×1.0°	0.94°×1.25°
18	SAM0-UNICON	Seoul National University, Seoul, Republic of Korea	1.25°×0.94°	1.1°×0.47°

520

Table 2. Classifications of well and poorly simulated groups in CMIP6 models.

No.	The poorly simulated group	No.	The well simulated group
1	ACCESS-CM2	5	CESM2-FV2
2	ACCESS-ESM1-5	6	CESM2-WACCM-FV2
3	CanESM5	10	MIROC6
9	CESM2	18	SAM0-UNICON

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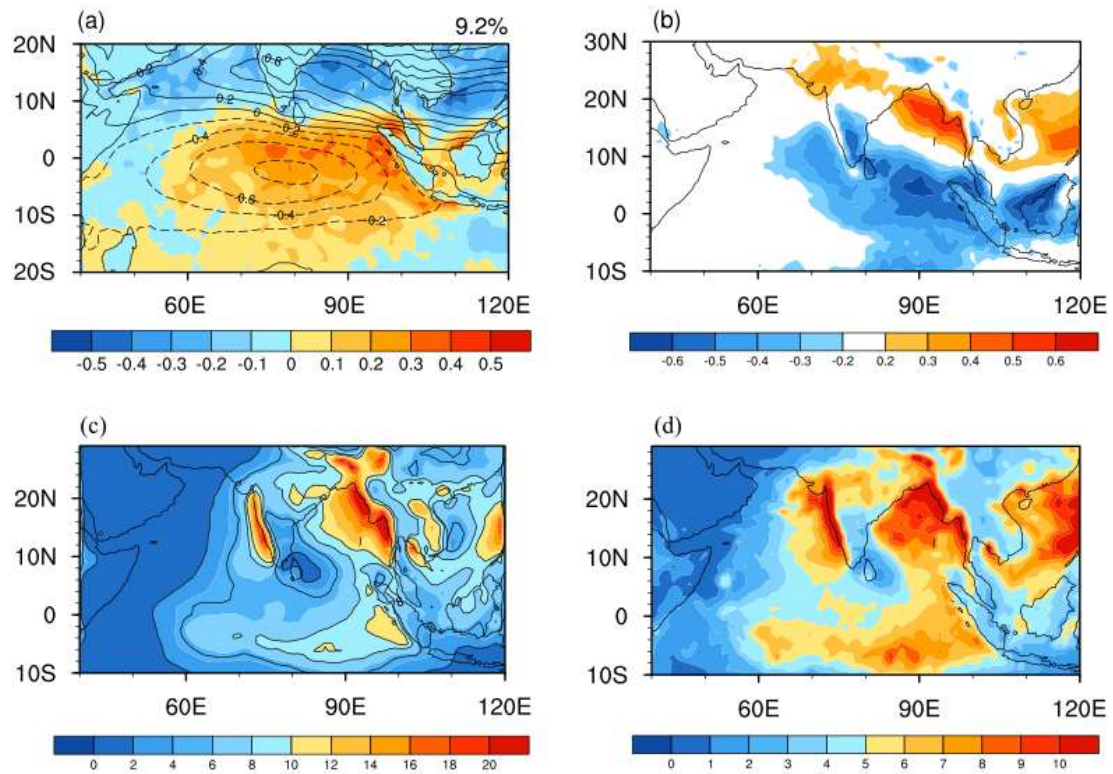
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Table 3. The numbers of CIO mode events in 18 CMIP6 air-sea coupled models.

No.	Model	Numbers of events	No.	Model	Numbers of events
1	ACCESS-CM2	57	10	MIROC6	56
2	ACCESS-ESM1-5	67	11	MPI-ESM-1-2-HAM	58
3	CanESM5	59	12	MPI-ESM1-2-HR	55
4	CESM2	60	13	MPI-ESM1-2-LR	56
5	CESM2-FV2	59	14	MRI-ESM2-0	60
6	CESM2-WACCM-FV2	64	15	NESM3	58
7	EC-Earth3-Veg	56	16	NorESM2-LM	61
8	GFDL-CM4	68	17	NorESM2-MM	57
9	IPSL-CM6A-LR	52	18	SAM0-UNICON	61

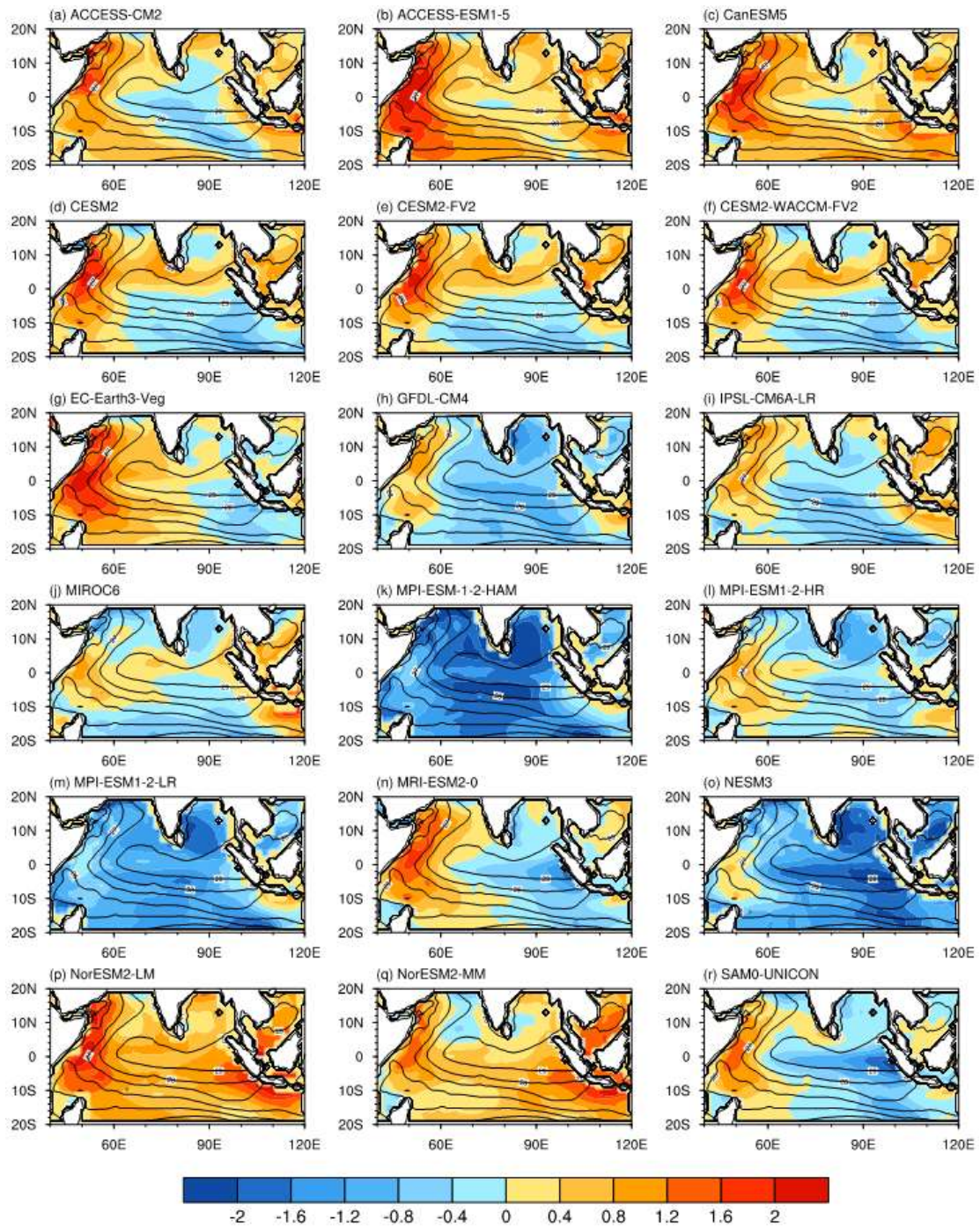
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528

529 **Figure 1** (a) The spatial structure of positive CIO mode obtained by daily OISST and
 530 ERA5 reanalysis from 1998 to 2014. Colors denote the SST node, and solid (dashed)
 531 contours denote westerly (easterly) winds. (b) Correlation maps of the CIO mode index
 532 with intraseasonal precipitation during the ISM. (c) Climatological mean of total
 533 precipitation (mm) calculated by TRMM from June to September during the period
 534 1998-2014. (d) is the same as (c), but for standard deviation of intraseasonal
 535 precipitation (mm).

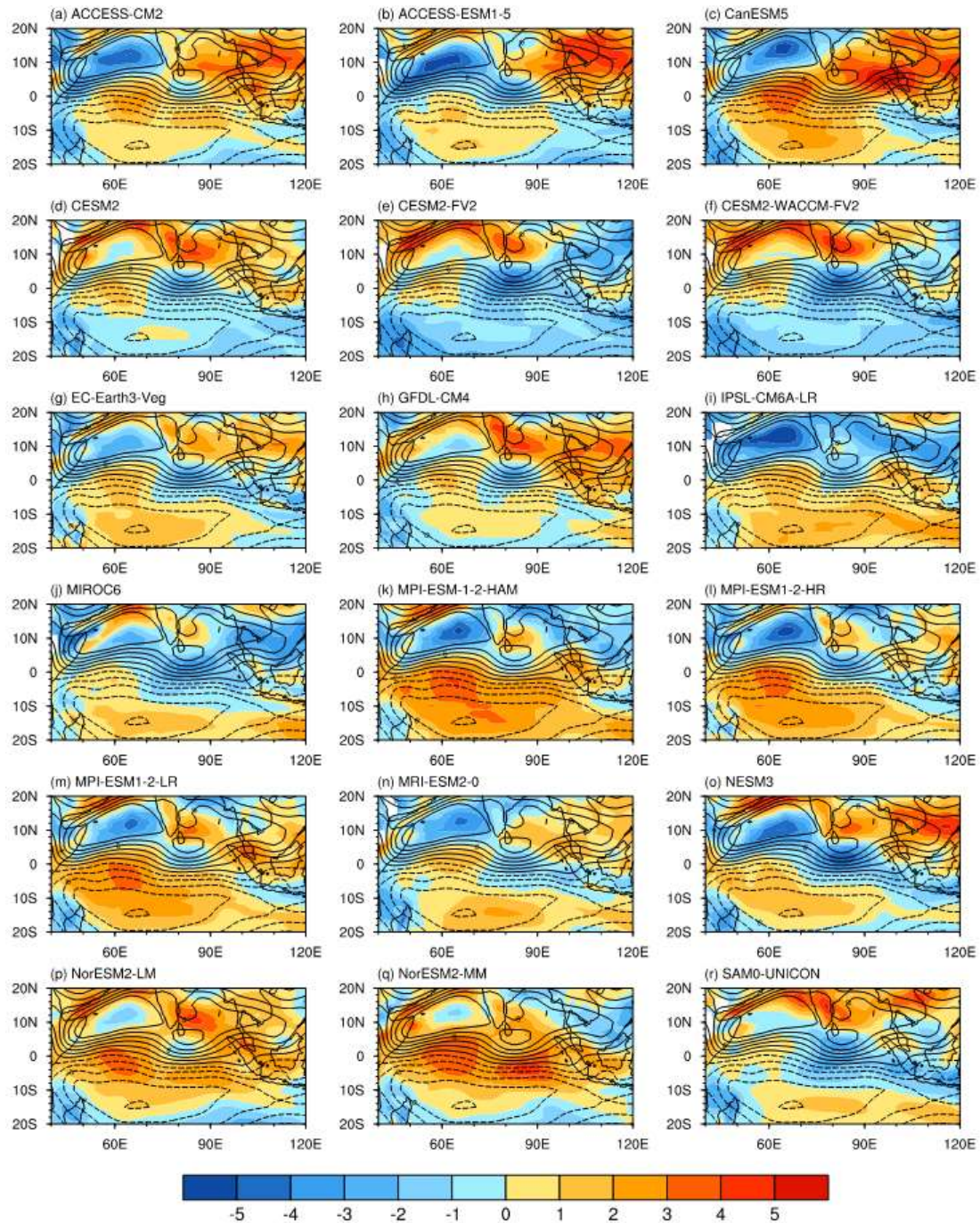
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538 **Figure 2** Differences of SST between observations and CMIP6 models (colors,
 539 simulations minus observations) averaged from June to September. Contours denote the
 540 observed SST. The unit is °C.

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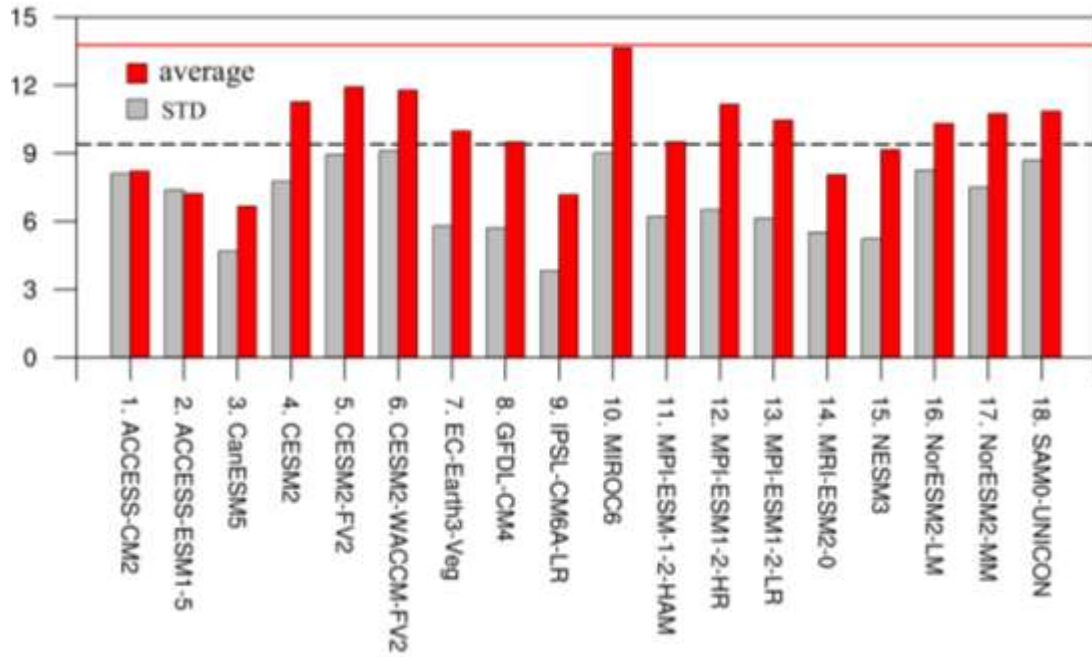


542

543 **Figure 3** Differences of U850 between observations and CMIP6 models (colors,
 544 simulations minus observations) averaged from June to September. Contours denote the
 545 observed U850. The unit is m s^{-1} .

546

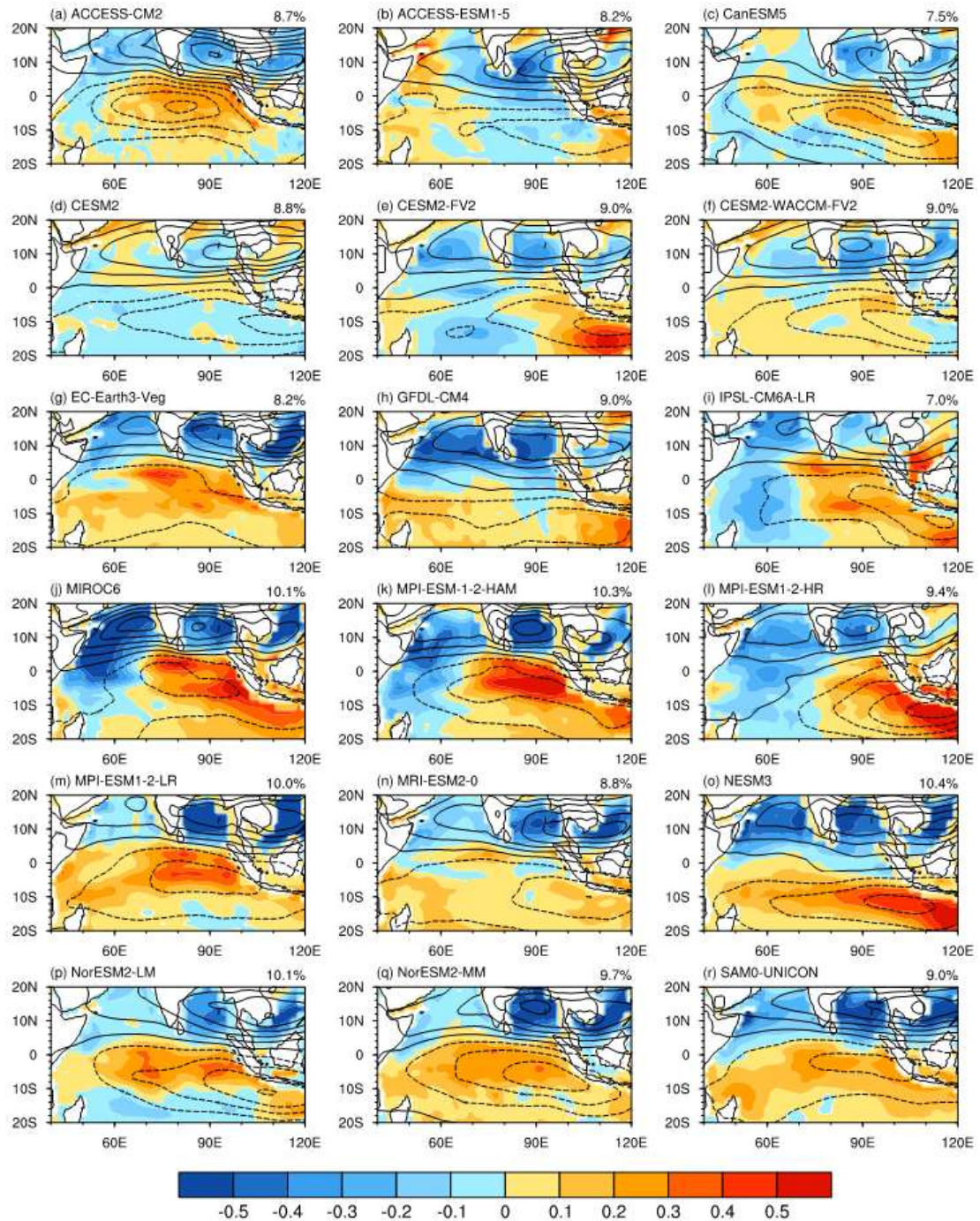
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548

549 **Figure 4** The total precipitation (red bars) and standard deviation of intraseasonal
 550 precipitation (gray bars) averaged in the BoB (10° - 20° N, 80° - 100° E, where the
 551 monsoonal precipitation is large) during boreal summer (June-September) in CMIP6
 552 models. The red and gray lines represent the observations.

553



554

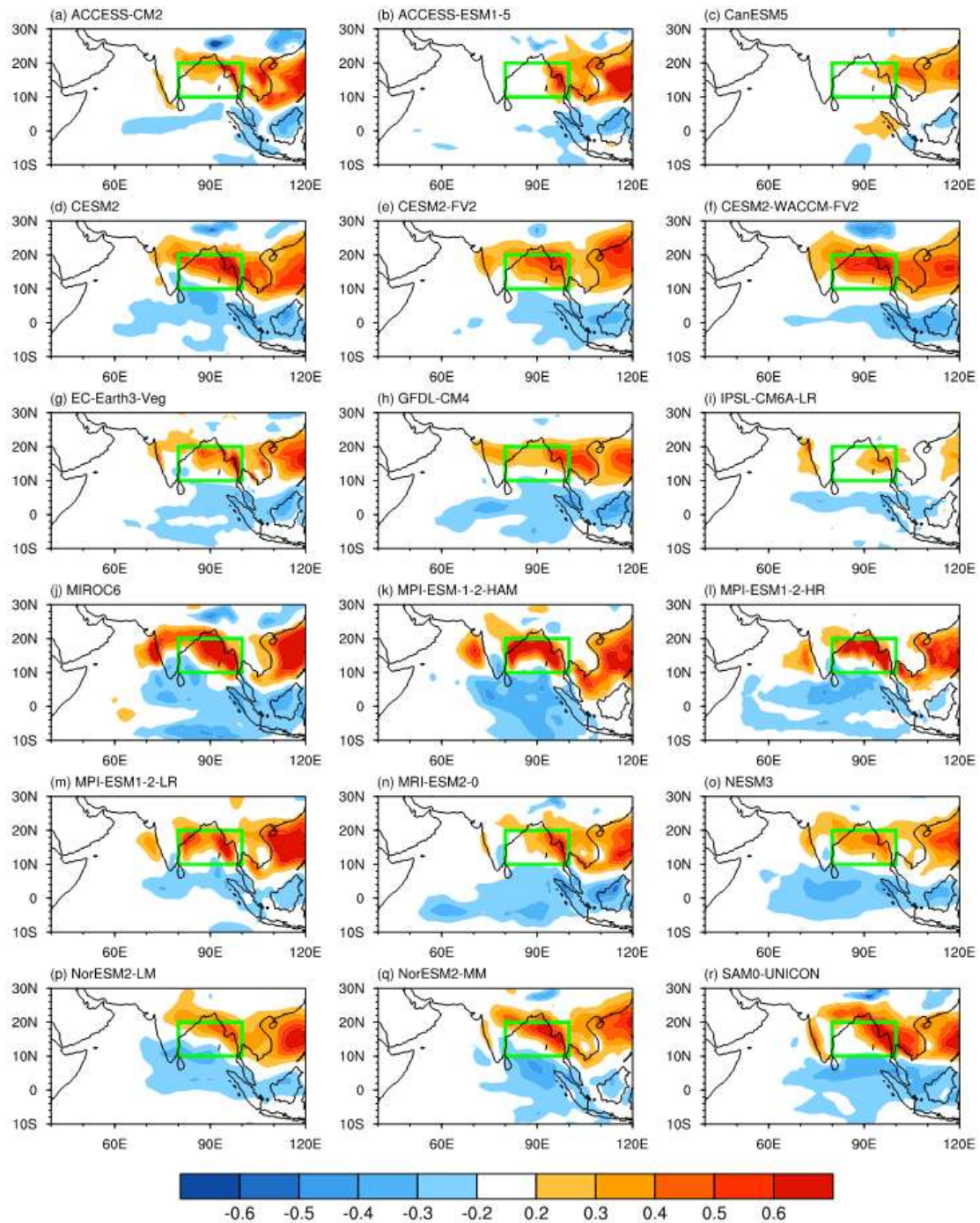
555 **Figure 5** The spatial structure of simulated CIO mode obtained by CMIP6 models.

556 Colors denote the SST node, and solid (dashed) contours denote westerly (easterly)

557 winds. The explained variance of simulated CIO mode is labelled in each model.

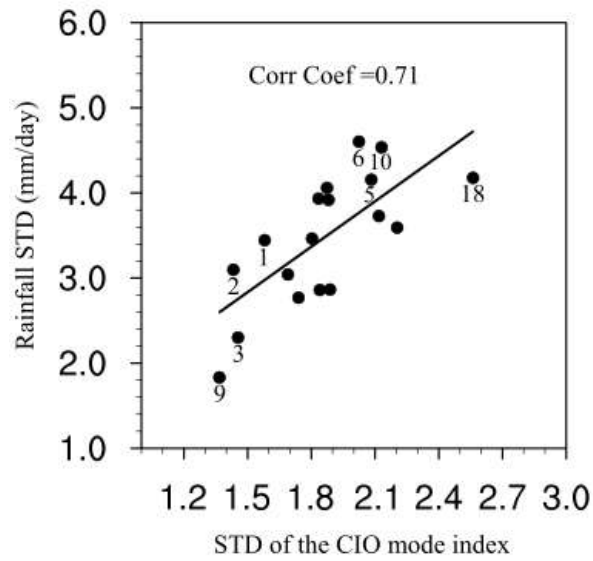
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560

561 **Figure 6** Correlation maps of the projected CIO mode index with the intraseasonal
 562 precipitation over the tropical Indian Ocean during the ISM in CMIP6 models. Green
 563 boxes represent 10°- 20°N, 80°-100°E, where the positive correlation coefficients are
 564 large in observation (Fig. 1b). Only correlations significant at a 95% confidence level
 565 are shown.

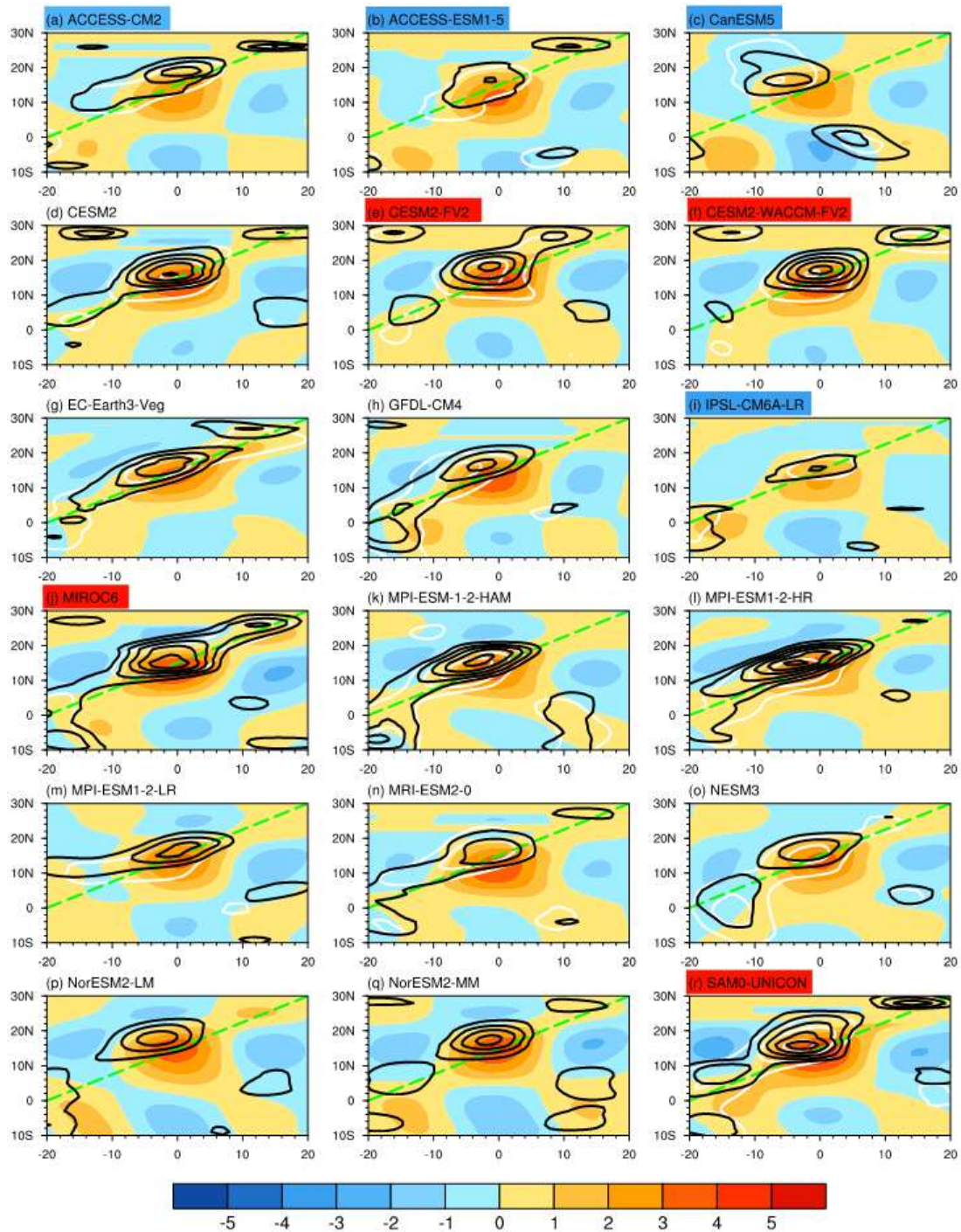


566

567 **Figure 7** The scatter plot of the standard deviation of intraseasonal precipitation (y-axis,
 568 mm day⁻¹) in the northern BoB (averaged within 10°N – 20°N and 85°E – 100°E) with
 569 respect to the standard deviation of the projected CIO mode indices (x-axis) in CMIP6
 570 models. The black line shows the linear regression of the scatter plot and the regression
 571 coefficient is statistically significant at the 99% confidence level.

572

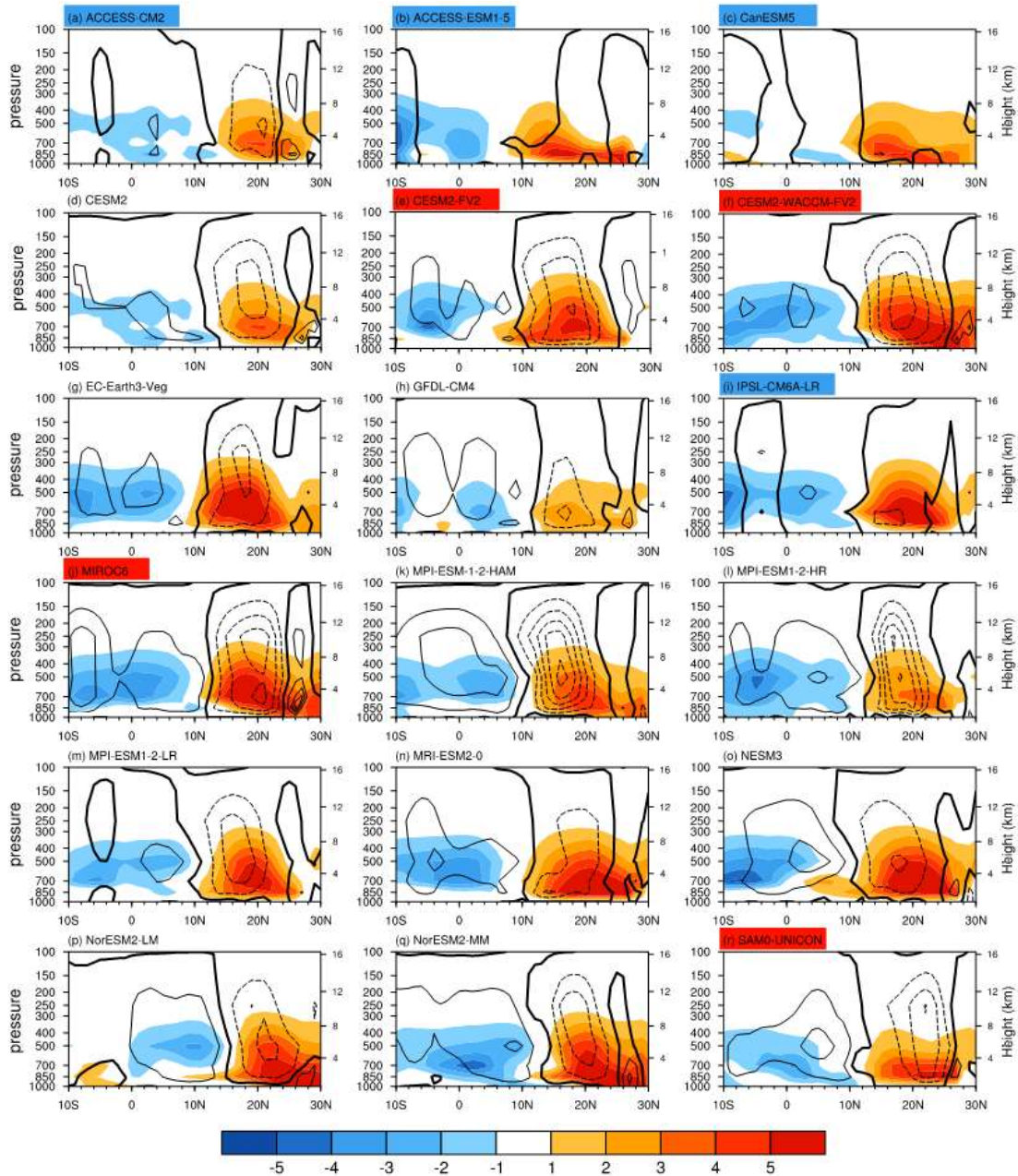
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575 **Figure 8** Composite Hovmöller diagram of intraseasonal precipitation (black solid
 576 contours; mm day^{-1}), intraseasonal OLR (white solid contours; W m^{-2}), and
 577 intraseasonal zonal wind (colors; m s^{-1}), averaged between 80°E and 90°E , calculated
 578 with CMIP6 models. Day 0 of the x-axis is the day when the projected CIO mode index
 579 reaches its maximum during the ISM. Negative days are before Day 0 and positive days
 580 are after Day 0. The green dashed lines represent the reference lines. The rainfalls from
 581 1 mm day^{-1} to 10 mm day^{-1} interval 1 mm day^{-1} are shown. The OLRs from -5 W m^{-2}

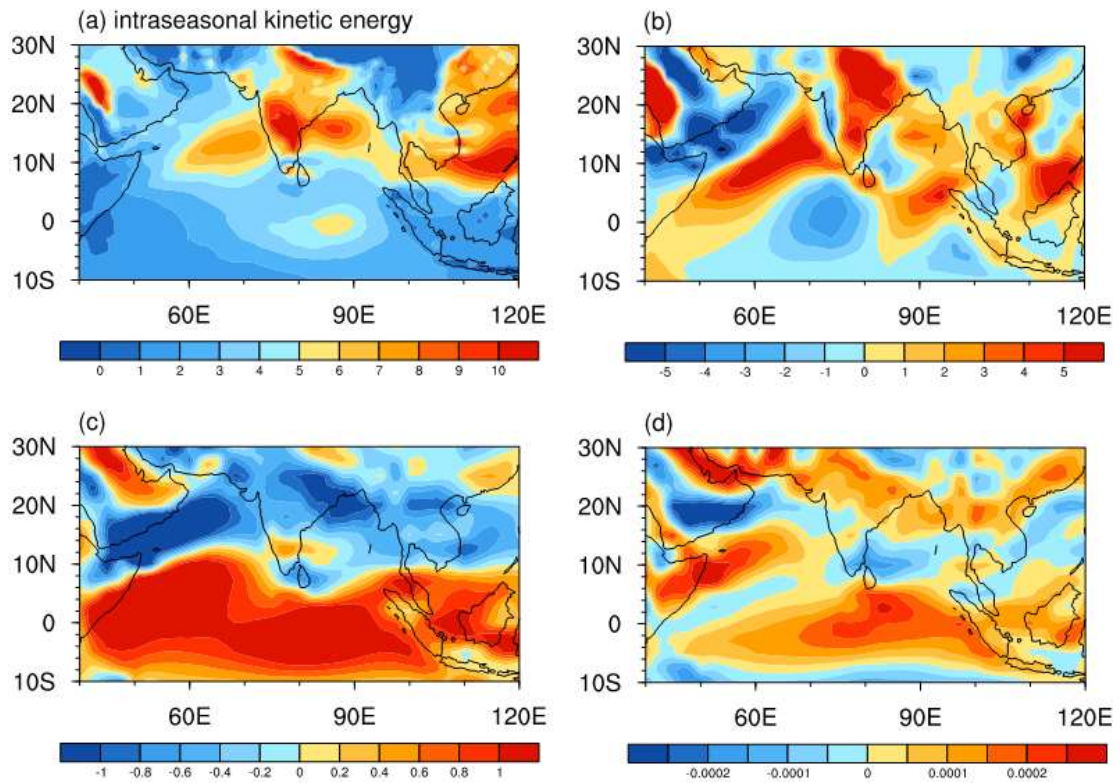
582 to -20 W m^{-2} interval 5 W m^{-2} are shown. The OLRs are not available in NorESM2-LM
583 and NorESM2-MM. The well (poorly) simulated group (listed in Tab.2) is shaded with
584 red (blue) colors.
585



586

587 **Figure 9** Composites of intraseasonal specific humidity (colors, g kg^{-1}) and omega
 588 (contours, Pa s^{-1}) between 80°E and 90°E for CMIP6 models when the projected CIO
 589 mode index reaches its maximum during the ISM. Only intraseasonal specific humidity
 590 significant at 95% confidence level are shown. Solid contours for negative omega
 591 (upstream) and dashed contours for positive omega (downstream). The omega from -5
 592 Pa s^{-1} to -2 Pa s^{-1} interval 1 Pa s^{-1} are shown. Lines with a value of zero are bolded. The
 593 well (poorly) simulated group (listed in Tab.2) is shaded with red (blue) colors.

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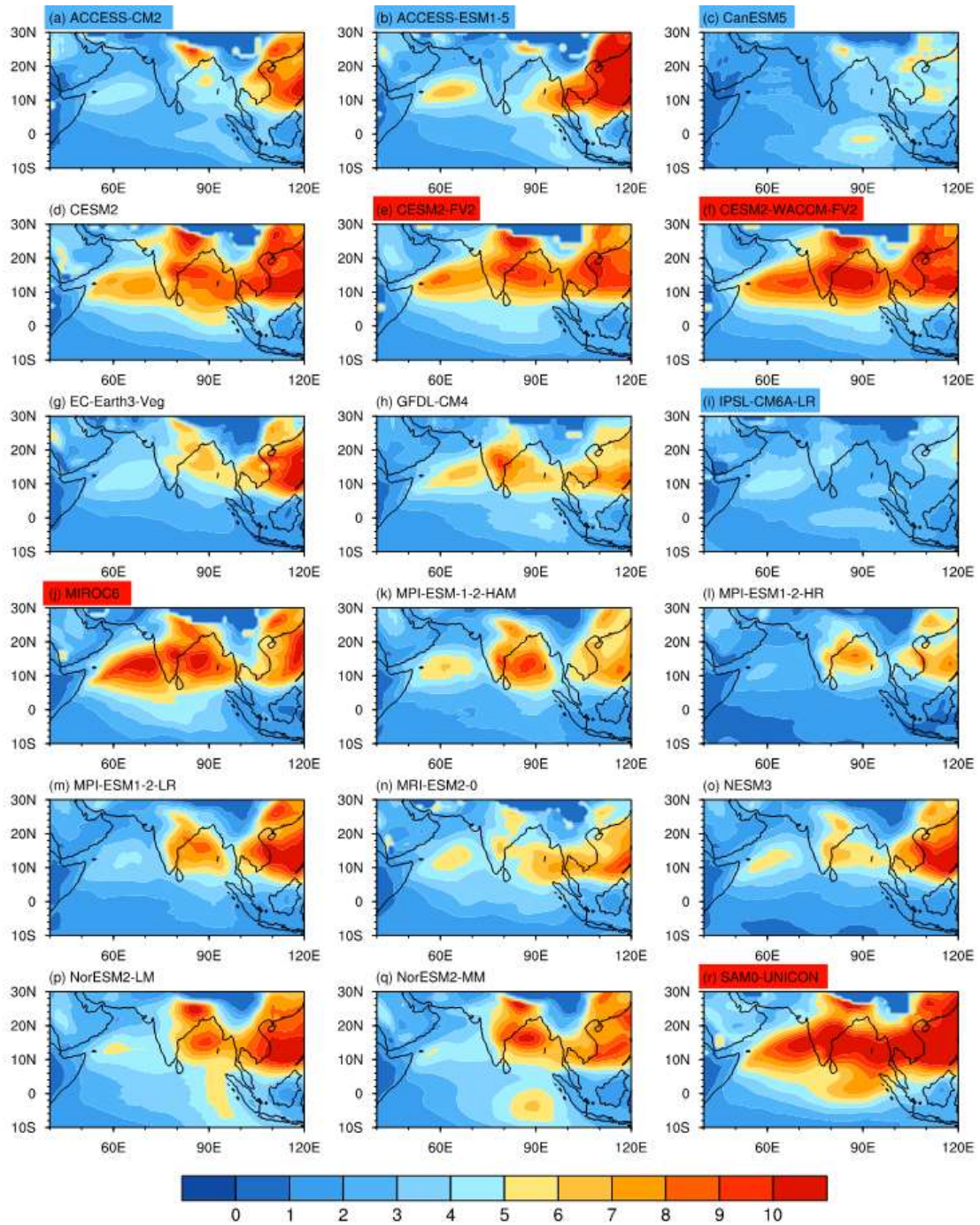
596 **Figure 10** (a) Intraseasonal kinetic energy (J kg^{-1}), (b) $[KE' \times \overline{KE}]$ ($\text{J day}^{-1} \text{kg}^{-1}$), (c)

597 $\partial \bar{u} / \partial y$ and (d) $\beta - \frac{\partial^2 \bar{u}}{\partial y^2}$ at 850 hPa averaged during boreal summer (from June to

598 September) in ERA5.

599

600

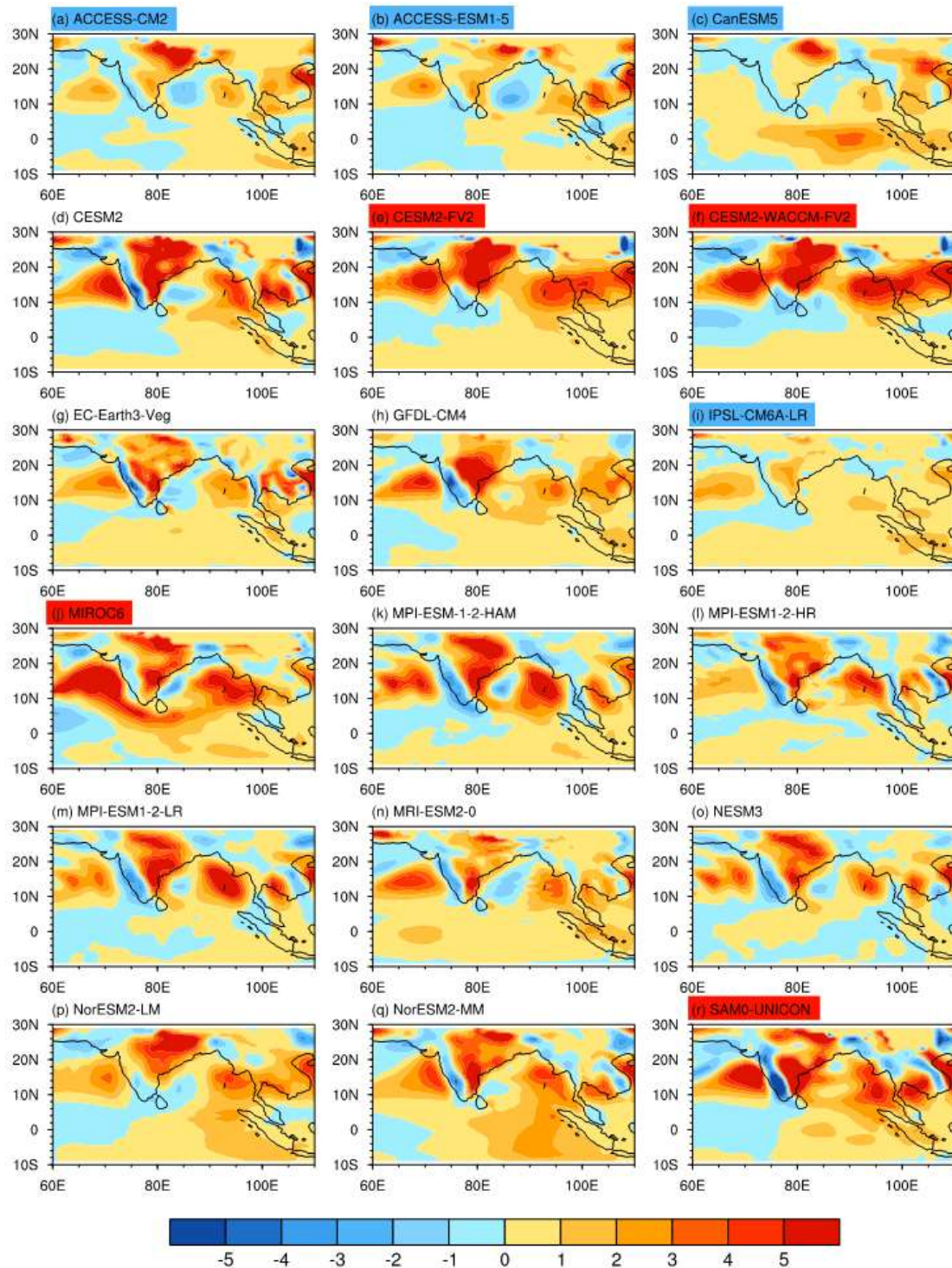


601

602 **Figure 11** is the same as Fig. 10a, but for CMIP6 models. The unit is J kg^{-1} . The well

603 (poorly) simulated group (listed in Tab.2) is shaded with red (blue) colors.

604



605

606 **Figure 12** is the same as Fig. 10b, but for CMIP6 models. The unit is $\text{J day}^{-1} \text{kg}^{-1}$. The

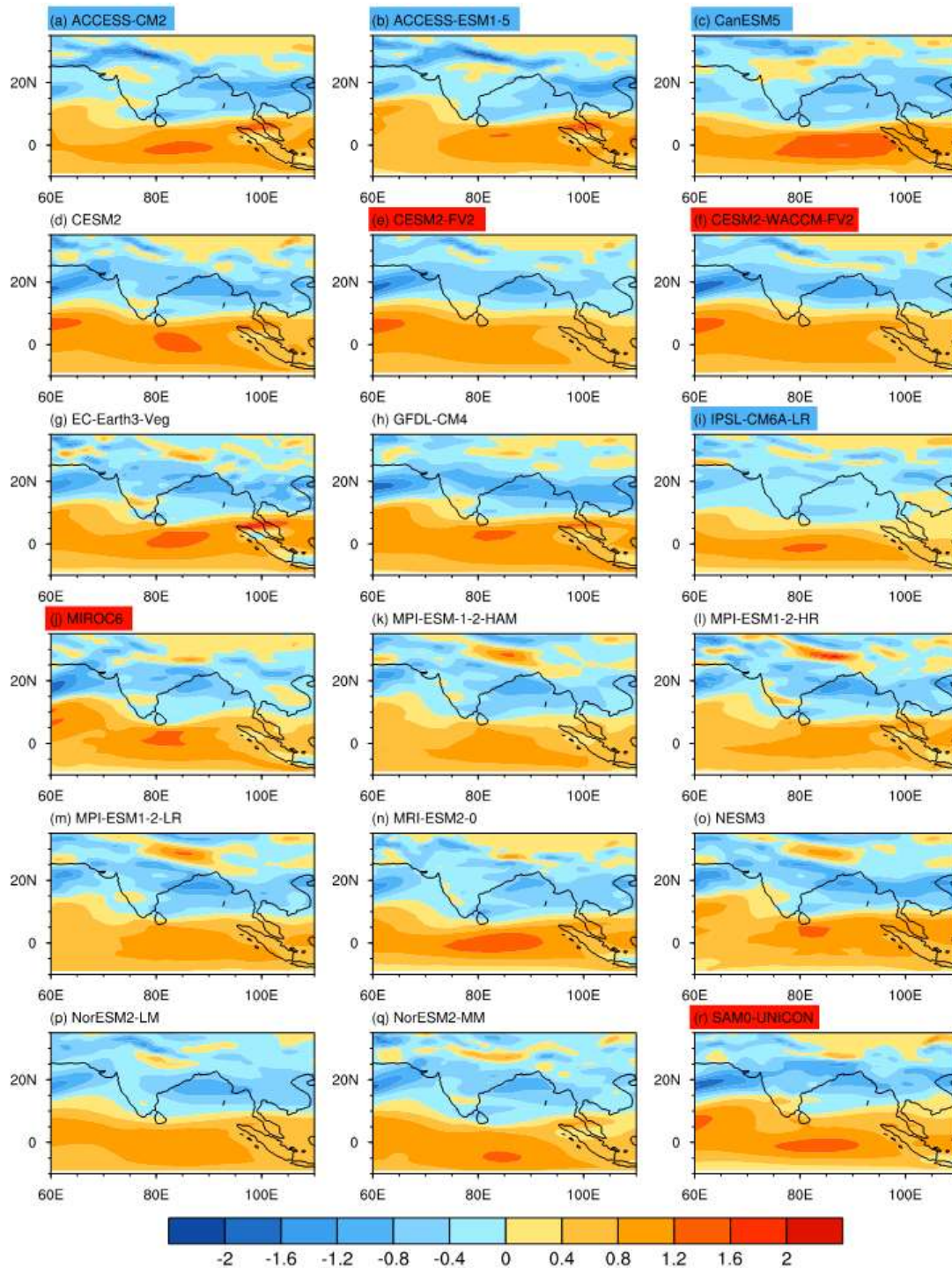
607 well (poorly) simulated group (listed in Tab.2) is shaded with red (blue) colors.

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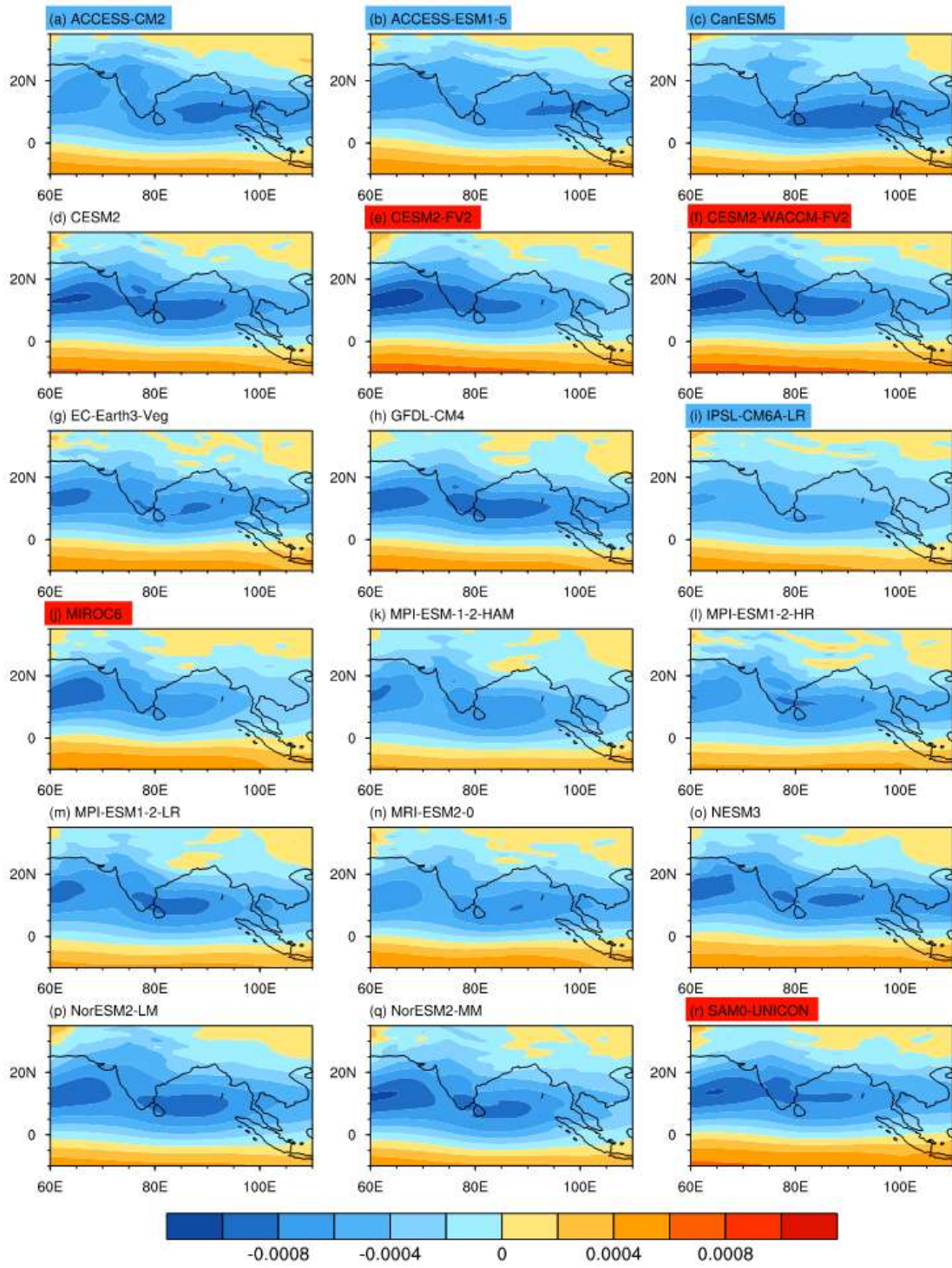


612

613 **Figure 13** is the same as Fig. 10c, but for CMIP6 models. The well (poorly) simulated

614 group (listed in Tab.2) is shaded with red (blue) colors.

615



616

617 **Figure 14** is the same as Fig. 10d, but for CMIP6 models. The well (poorly) simulated
 618 group (listed in Tab.2) is shaded with red (blue) colors.